


DATA MINING CLASSIFICATION TECHNIQUES AND PERFORMANCES ON MEDICAL DATA

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ABSTRACT

This study evaluates the performance of classification techniques with the application of several software, among them are Rosetta, Tanagra, Weka and Orange. The classification technique has been tested on six medical datasets from the UCI Machine Learning Repository. The study will help researchers to select the best suitable technique of classification problem for medical datasets in term of classification accuracy. In this thesis, sixteen classification techniques have been evaluated and compared. These are Radial Basis Function (RBF), Multilayer Perceptron (MLP) Neural Networks, Multi Linear Regression (MLR), Logistic Regression (LR), Classification Tree (ID3, C4.5, J48, CART), Naive Bayes (NB), Support Vector Machines (SVM), k- Nearest Neighbors (kNN), Linear discriminate analysis (LDA), Rule based classifier, Standard voting, Voting with object tracking and Standard/ tuned voting (RSES). The experiments have been validated using 10-fold cross validation method. The results of the study shows that the most suitable classification technique is NB with an average classification accuracy of 90.13% and an average error rate of 9.87%. The worst classification technique is SLR with an average classification accuracy of 50.16% and an average error rate of 49.84%. The classification techniques has been ranked from the best to the worst based on average classification accuracy and average error rate. The top of the rank is NB and the bottom is SLR. The sequence of ranking from the best to the worst is NB, LDA, LR, SVM, C4.5, MLP, RBF, kNN, RuleB, ID3, CART, J48, SV, RSES, V, and SLR.

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CHAPTER 1 *

INTRODUCTION

1.1 Overview

With the enormous amount of data stored in files, databases, and other repositories, it is increasingly important to develop powerful means for analysis, interpretation and extraction of interesting knowledge that could help in decision-making. Data Mining, also popularly known as Knowledge Discovery in Databases (KDD), refers to the nontrivial extraction of implicit, previously unknown and potentially useful information from data in databases. It uses machine learning, statistical and visualization techniques to discover and present knowledge in a form which is easily comprehensible to humans (Frawley *et al.*, 1992).

The kinds of patterns that can be discovered depend upon the data mining tasks employed. Generally, there are two types of data mining tasks: descriptive data mining tasks that describe the general properties of the existing data, and predictive data mining tasks that attempt to do predictions based on inference on available data. There are many ways to approach data mining problems, including creating statistical models, classification, predictive modeling, clustering, finding association rules and sequence analysis, and anomaly detection. Classification is one of the most important tasks in data mining where its main

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